Data-Driven Decision-Making In Enterprise Applications

Linear Programming

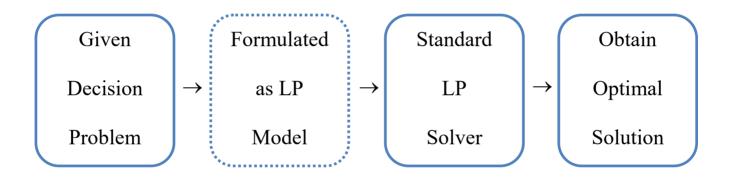
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April 30, 2020

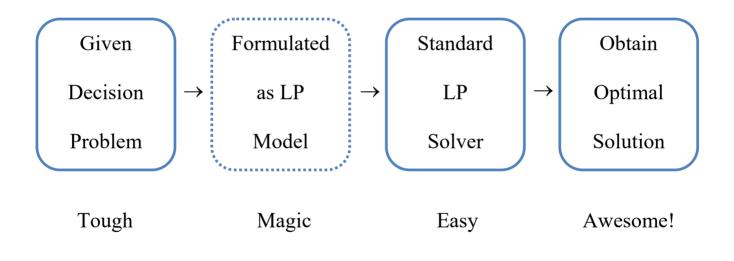


Decision-Making Using Linear Programming





Decision-Making Using Linear Programming



Linear Programming & Focus

- What is a Linear Program?
- Theoretical Foundations
- Standard Solution Algorithms
- Tricks to formulate decision problems as LP
- Examples, Examples, Examples, . . .

Data-Driven Decision-Making in Enterprise Applications – Linear Programming

HPI

What is a Linear Program?

Decision variables x_1, x_2, \dots

The controls to be determined.

Constraints via ≤,=,≥

Expressed linearly in x_1, x_2, \dots .

One objective: max or min

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Expressed linearly in x_1, x_2, \dots .

- Does the solver help you to define the right LP?
- How to formulate a given problem as an LP?
- What does "expressed linearly" means?



Which Terms are Linear in the Variables $x_1, x_2, ... ?$

$$3x_{1} - 2x_{2}, \quad x_{1}^{-3}, \quad a \cdot \ln(x_{1}), \quad \ln(a) \cdot x_{1}$$

$$a \cdot x_{1}, \quad a \cdot b \cdot x_{1}, \quad a \cdot x_{1}^{2}, \quad \sqrt{\ln(a)} \cdot \frac{x_{1} + x_{2}}{b} \cdot \sin(a^{2})$$

$$|x_{1}|, \quad \max(x_{1}, 5), \quad x_{1}^{2} / x_{1}, \quad (x_{1} - 3) \cdot (x_{2} + 3), \quad a^{2} \cdot x_{1},$$

$$x_{1} \cdot x_{2}, \quad x_{1} / x_{2}, \quad 1_{\{x_{1} = 5\}} := \text{ if } x_{1} = 5 \text{ then } 1 \text{ else } 0$$



$$3x_{1} - 2x_{2}, \quad x_{1}^{-3}, \quad a \cdot \ln(x_{1}), \quad \ln(a) \cdot x_{1}$$

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$$x_{1} \cdot x_{2}, \quad x_{1} / x_{2}, \quad 1_{\{x_{1} = 5\}} := \text{ if } x_{1} = 5 \text{ then } 1 \text{ else } 0$$

Example of a Linear Program

Objective:

$$\max_{x_1, x_2 \in \mathbb{R}} 2 \cdot x_1 + 3 \cdot x_2$$

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Constraints: $0.5 \cdot x_1 \le 4 - x_2$,

$$0 \le x_1 \le 3, \quad x_2 \ge 0$$

Example of a Linear Program

Objective: $\max_{x_1, x_2 \in \mathbb{R}} 2 \cdot x_1 + 3 \cdot x_2$

Constraints: $0.5 \cdot x_1 \le 4 - x_2$,

$$0 \le x_1 \le 3, \quad x_2 \ge 0$$

- Only use linear expressions
- Only use \leq , =, \geq in the constraints (<, \neq , > are not allowed!)
- Is such an LP understandable for all solvers?



Example of a Linear Program in Standard Form

Objective: $\max_{x_1, x_2 \ge 0} 2 \cdot x_1 + 3 \cdot x_2$

subject to $0.5 \cdot x_1 + 1 \cdot x_2 \le 4$

$$1 \cdot x_1 + 0 \cdot x_2 \le 3$$

Standard form:

- Linear combinations of (*non-negative*) variables
- Use \leq (or =) in all constraints
- No variables on the "*right-hand side*" of the constraints



Example of a Linear Program in Standard Form

Objective: $\max_{x_1, x_2 \ge 0} 2 \cdot x_1 + 3 \cdot x_2$ s.t. $0.5 \cdot x_1 + 1 \cdot x_2 \le 4$

$$\max_{x_1, x_2 \ge 0} \vec{c} \cdot \vec{x} \quad \text{s.t.} \quad A \cdot \vec{x} \le \vec{b}$$

with $\vec{c} \coloneqq (2,3), \quad \vec{b} \coloneqq (4,3),$
$$A \coloneqq \begin{pmatrix} 0.5 & 1\\ 1 & 0 \end{pmatrix}$$

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 $1 \cdot x_1 + 0 \cdot x_2 \le 3$



Example of a Linear Program in Standard Form

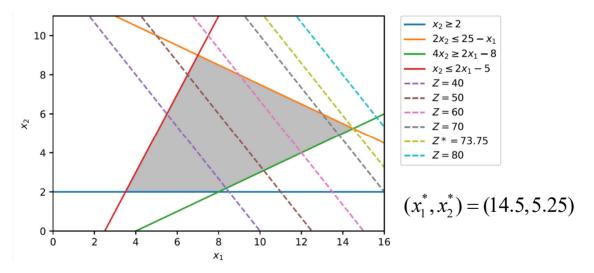
Objective:
$$\max_{x_1, x_2 \ge 0} 2 \cdot x_1 + 3 \cdot x_2$$
$$\max_{x_1, x_2 \ge 0} \vec{c} \cdot \vec{x} \quad \text{s.t.} \quad A \cdot \vec{x} \le \vec{b}$$
with $\vec{c} \coloneqq (2,3), \quad \vec{b} \coloneqq (4,3),$
$$1 \cdot x_1 + 0 \cdot x_2 \le 3$$
$$A \coloneqq \begin{pmatrix} 0.5 & 1 \\ 1 & 0 \end{pmatrix}$$

Optimal Solution: Computed by solvers using standard algorithms For further reading see: Simplex Algorithm, Convex Polyeder Theory, etc.

Graphical Solutions (Continuous LP)

Example:

$$\max_{x_1 \ge 0, x_2 \ge 2} 4 \cdot x_1 + 3 \cdot x_2 \quad \text{s.t.} \quad 2 \cdot x_2 \le 25 - x_1, \ 4 \cdot x_2 \ge 2x_1 - 8, \ x_2 \le 2x_1 - 5$$

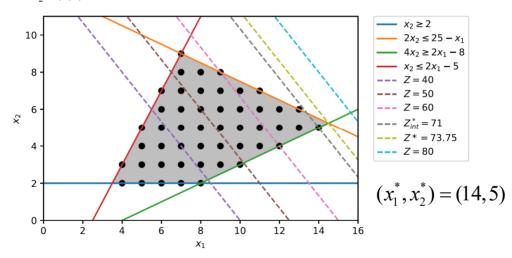


• Continuous LP: One corner of the feasible space is always optimal!

Graphical Solutions (Integer LP)

Example:

 $\max_{\substack{x_1=0,1,2,\dots\\x_2=2,3,4,\dots}} 4 \cdot x_1 + 3 \cdot x_2 \quad \text{s.t.} \quad 2 \cdot x_2 \le 25 - x_1, \ 4 \cdot x_2 \ge 2x_1 - 8, \ x_2 \le 2x_1 - 5$



• Integer LP: Use discrete variables $x_1, x_2, ... \in \{0,1\}$ or $\in \{0,1,2,...\}$ Optimal solutions via integer solvers (see "Branch & Bound")

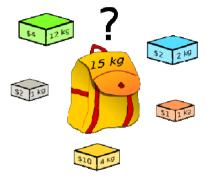
Examples, Examples, Examples

- I Knapsack Problem
- II Magic Square
- III Matrix Inversion
- IV Project Assignment Problems
- V Mixed Equilibria (Game Theory)

I Knapsack Problem

Given parameters:

- *N* Number of potential items
- u_i Utility of packing item *i*, *i*=1,...,*N*
- S_i Space of item *i*, *i*=1,...,*N*
- *C* Capacity of the knapsack
- Problem: Decide which items to take to maximize total utility, while not exceeding the knapsack's capacity





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I LP Model for the Knapsack Problem

Variables: x_i binary variable whether to pack item *i*, *i*=1,...,N

LP:
$$\max_{x_1,...,x_N \in \{0,1\}} \sum_{i=1,...,N} u_i \cdot x_i$$

maximize total utility

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I LP Model for the Knapsack Problem

< C

Variables: x_i binary variable whether to pack item *i*, *i*=1,...,N

LP:
$$\max_{x_1,...,x_N \in \{0,1\}} \sum_{i=1,...,N} u_i \cdot x_i$$

maximize total utility

s.t.
$$\sum_{i=1,\dots,N} s_i \cdot x_i$$

satisfy budget constraint

Let the solver do the rest!

Done.



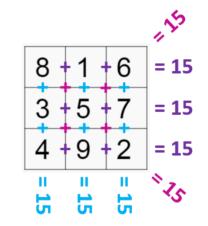
I Implementation of the Knapsack LP (AMPL)

$\max_{x_1,,x_N \in \{0,1\}} \sum_{i=1,,N} u_i \cdot x_i \qquad \text{s.t.} \qquad \sum_{i=1,,N} s_i$	• 7	$x_i \leq C$
reset;		
<pre>param C := Uniform(50,100);</pre>	#	capacity
<pre>param N := Uniform(100,200);</pre>	#	number of items
<pre>param u {i in 1N} := Uniform(3,8)</pre>	#	utility of item i
<pre>param s {i in 1N} := Uniform(1,2)</pre>	#	space of item i
<pre>var x {i in 1N} binary;</pre>	#	binary variables
<pre>maximize LP: sum{i in 1N} u[i]*x[i];</pre>	#	objective
<pre>subject to NB: sum{i in 1N} s[i]*x[i] <</pre>	<=	C; # budget con.
solve; display x;	#	obtain solution

II Magic Square

 $N = n^2$ Numbers i=1,...,N to be filled in a square of size $n \times n$, $n \ge 3$

Problem: Assign the numbers i=1,...,Nto the cells of a square such that the sum of each row and each column is the same.

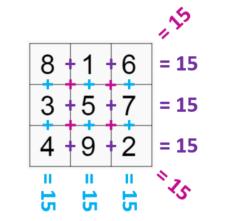


Task: For n=3 the sum is 15. What is the magic number for n=4?

II Magic Square

 $N = n^2$ Numbers i=1,...,N to be filled in a square of size $n \times n$, $n \ge 3$

Problem: Assign the numbers i=1,...,Nto the cells of a square such that the sum of each row and each column is the same.



Note: The magic number is $C := N \cdot (N+1) / 2 / n = n \cdot (n^2 + 1) / 2 = 15$ = 34

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n=3

II LP Model for the Magic Square

Variables: $x_{i,j,k}$ binary variable whether number k, k=1,...,Nis assigned to row i and column j, i,j=1,...,n HPI

II LP Model for the Magic Square

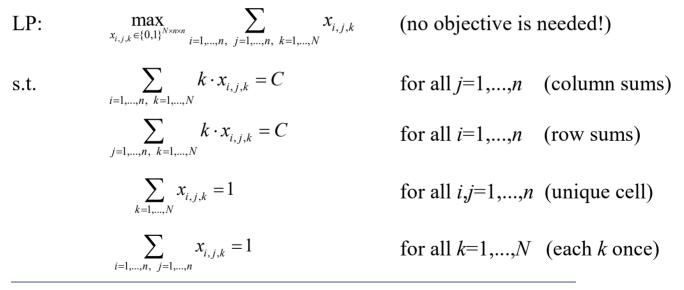
Variables: $X_{i,j,k}$ binary variable whether number k, k=1,...,Nis assigned to row *i* and column *j*, *i,j*=1,...,*n*

LP:
$$\max_{x_{i,j,k} \in \{0,1\}^{N \times n \times n}} \sum_{i=1,\dots,n, j=1,\dots,N} x_{i,j,k} \quad \text{(no objective is needed!)}$$

s.t.
$$\sum_{i=1,\dots,n, k=1,\dots,N} k \cdot x_{i,j,k} = C \quad \text{for all } j=1,\dots,n$$
$$\sum_{j=1,\dots,n, k=1,\dots,N} k \cdot x_{i,j,k} = C \quad \text{for all } i=1,\dots,n$$

II LP Model for the Magic Square

Variables: $x_{i,j,k}$ binary variable whether number k, k=1,...,Nis assigned to row i and column j, i,j=1,...,n

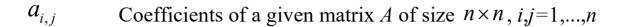


Hints

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- At first: **Think**! Do not start to implement too early
- Do not hope that the LP/solver thinks in your favour!
- Remember: You have to **force** the solver to do the right thing
- Try as hard as you can to formulate the problem **linearly**
- Do not hesitate to use many variables and constraints
- 95% of the work is setting up the LP, 5% is implementation!

III Matrix Inversion



Problem: Find the **inverse matrix** A^{-1} of matrix Adetermined by its coefficients $x_{i,j}$, i,j=1,...,n.

Note: The inverse A^{-1} is uniquely determined by satisfying:

$$A^{-1} \times A = A \times A^{-1} = I \coloneqq \begin{bmatrix} 1 & 0 \\ & \dots \\ 0 & 1 \end{bmatrix}$$



III LP Model for Matrix Inversions

Variables: $x_{i,j}$ continuous variable for the coefficient of row *i* and column *j* of the inverse matrix A^{-1} , *i,j*=1,...,*n*

LP:

III LP Model for Matrix Inversions

Variables: $X_{i,j}$ continuous variable for the coefficient of
row *i* and column *j* of the inverse matrix A^{-1} , i,j=1,...,nLP: $\max_{x_{i,j} \in \mathbb{R}^{n \times n}} \sum_{i,j=1,...,n} x_{i,j}$ (objective is optional!)s.t. $\sum_{k=1,...,n} a_{i,k} \cdot x_{k,j} = 1_{\{i=j\}}$ for all i,j=1,...,n $(A \cdot X = I)$

III Implementation of the Matrix Inversion LP

Variables: $X_{i,i}$ continuous variable for the coefficient of row *i* and column *j* of the inverse matrix A^{-1} , *i*,*j*=1,...,*n* $\max_{x_{i,j} \in \mathbb{R}^{n \times n}} \sum_{i, j=1,\dots,n} x_{i,j}$ LP: (objective is optional!) $\sum a_{i,k} \cdot x_{k,j} = \mathbf{1}_{\{i=j\}}$ for all i, j=1, ..., n $(A \cdot X = I)$ s.t. param n := 5;# size of A param a {i in 1..n,j in 1..n} := Uniform(-1,1); # coeff. of A # coeff. of A^{-1} x {i in 1..., j in 1..., }; var subject to NB{i in 1..n,j in 1..n}: # identity $sum\{k \text{ in } 1..n\} a[i,k]*x[k,j] = if i=j \text{ then } 1 \text{ else } 0;$ solve; display x; # solution output

IV Project Assignment Problem

Given parameters:

- N Number of workers/projects
- $W_{i,j}$ Willingness of worker i=1,...,N

to take project j=1,...,N



Problem: Decide how to distribute all projects

in order to maximize total welfare,

while assigning one project to each worker

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IV LP Model for the Project Assignment Problem

Variables: $x_{i,j}$ binary variable whether project *i*, *i*=1,...,*N* is assigned to worker *j*, *j*=1,...,*N*

LP:

IV LP Model for the Project Assignment Problem

Variables: $x_{i,j}$ binary variable whether project *i*, *i*=1,...,*N* is assigned to worker *j*, *j*=1,...,*N*

LP:
$$\max_{x_{i,j} \in \{0,1\}^{N \times N}} \sum_{i=1,...,N} w_{i,j} \cdot x_{i,j}$$

s.t.
$$\sum_{i=1,...,N} x_{i,j} = 1 \quad \text{for all } j=1,...,N \quad \text{(each worker gets 1 project)}$$
$$\sum_{j=1,...,N} x_{i,j} = 1 \quad \text{for all } i=1,...,N \quad \text{(each project is assigned)}$$

Next Week



Homework: Check out AMPL's student version

https://ampl.com/try-ampl/download-a-free-demo/

Review the Examples

Outlook:

- More Examples, e.g., Mixed Strategy Equilibria (Game Theory)
- Tricks to linearize Nonlinearities
- Implementations and Solvers
- Multi-objective/Penalty Approaches

Overview

Week	Dates	Торіс	
1	April 27/30	Introduction + Linear Programming	
2	May 4 / (7)	Linear Programming II	
3	May 11/14	Exercise Implementations	
4	May 18	Linear + Logistic Regression	(Thu May 21 "Himmelfahrt")
5	May 25/28	Dynamic Programming	(Mon June 1 "Pfingstmontag")
6	June 4	Dynamic Pricing Competition	
7	June 8/11	Project Assignments	
8	June 15/18	Robust + Nonlinear Optimization	
9	June 22/25	Work on Projects: Input/Support	
10	June 29/2	Work on Projects: Input/Support	
11	July 6/9	Work on Projects: Input/Support	
12	July 13/16	Work on Projects: Input/Support	
13	July/Aug	Finish Documentation (Deadline: Au	ug 31)